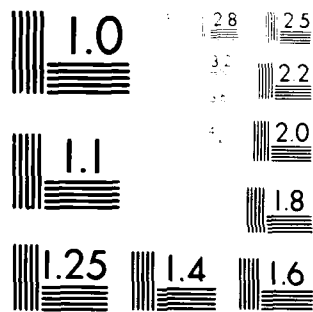


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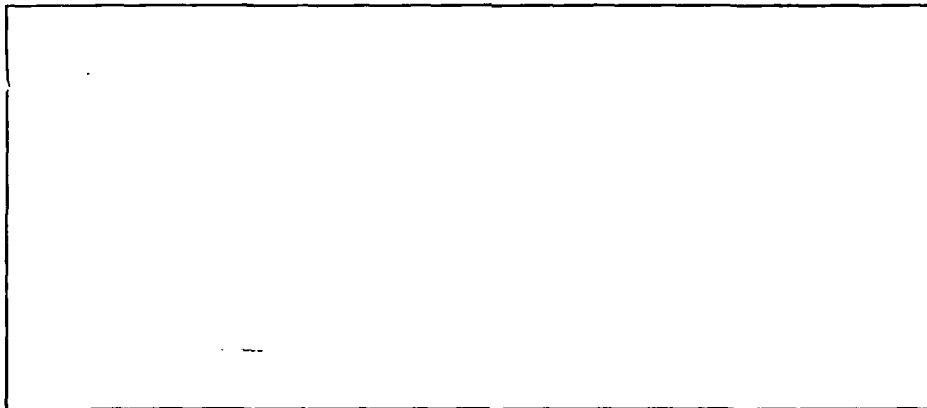
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**Abstract**

The preconditioned conjugate gradient (PCG) method is an effective means for solving systems of linear equations where the coefficient matrix is symmetric and positive definite. The incomplete LDL<sup>t</sup> factorizations are a widely used class of preconditionings, including the SSOR, Dupont-Kendall-Rachford, Generalized SSOR, ICCG(0), and MICCG(0) preconditionings. The efficient implementation of PCG with a preconditioning from this class is discussed.

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Efficient Implementation of a Class  
of Preconditioned Conjugate Gradient Methods

(1) Stanley C. Eisenstat

(7) Research Report #185

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## 1. Introduction

Consider the system of  $N$  linear equations

$$(1) \quad A x = b,$$

where the coefficient matrix  $A$  is symmetric and positive definite. When  $A$  is large and sparse, the preconditioned conjugate gradient (PCG) method is an effective means for solving (1) [2, 4, 5, 9, 13]. Given an initial guess  $x_0$ , we generate a sequence  $\{x_k\}$  of approximations to the solution  $x$  as follows:

$$(2a) \quad p_0 = r_0 = b - Ax_0$$

$$(2b) \quad \text{Solve } Mr'_0 = r_0$$

FOR  $k = 0$  STEP 1 UNTIL Convergence DO

$$(2c) \quad a_k = (r_k, r'_k) / (p_k, Ap_k)$$

$$(2d) \quad x_{k+1} = x_k + a_k p_k$$

$$(2e) \quad r_{k+1} = r_k - a_k Ap_k$$

$$(2f) \quad \text{Solve } Mr'_{k+1} = r_{k+1}$$

$$(2g) \quad b_k = (r_{k+1}, r'_{k+1}) / (r_k, r'_k)$$

$$(2h) \quad p_{k+1} = r'_{k+1} + b_k p_k$$

The effect of the preconditioning matrix  $M$  is to increase the rate of convergence of the basic conjugate gradient method of Hestenes and Stiefel [11]. The number of multiply-adds per iteration is just  $5N$ , plus the number required to form  $Ap_k$ , plus the number required to solve  $Mr'_k = r_k$ .

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One widely used class of preconditionings are the incomplete  $LDL^t$  factorizations

$$(3) \quad M = (\tilde{D}+L) \tilde{D}^{-1} (\tilde{D}+L)^t,$$

where  $A = L+D+L^t$ ,  $L$  is strictly lower triangular, and  $D$  and  $\tilde{D}$  are positive diagonal. This class includes the SSOR [9], Dupont-Kendall-Rachford [7], Generalized SSOR [1], ICCG(0) [13], and MICCG(0) [10] preconditionings. Letting  $NZ(A)$  denote the number of nonzero entries in the matrix  $A$ , a straight-forward implementation of PCG with a preconditioning from this class<sup>1</sup> would require  $6N+2NZ(A)$  multiply-adds per iteration.<sup>2</sup>

In this brief note, we show how to reduce the work to  $8N+NZ(A)$  multiply-adds, asymptotically half as many as the straight-forward implementation.<sup>3</sup> We give details in Section 2, and consider some generalizations in Section 3.

## 2. Implementation

The linear system (1) can be restated in the form

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<sup>1</sup> Writing  $M$  as  $(\tilde{D}+L)(I+\tilde{D}^{-1}L^t)$ , we solve  $Mr'_k = r_k$  by solving the triangular systems  $(\tilde{D}+L)t_k = r_k$ ,  $(I+\tilde{D}^{-1}L^t)r'_k = t_k$ .

<sup>2</sup>  $2N$  (respectively,  $N$ ) multiply-adds can be saved by symmetrically scaling the problem to make  $\tilde{D} = I$  (respectively,  $D = I$ ).

<sup>3</sup> A similar speedup for pairs of linear iterative methods is given in [6].

$$(4) \quad [(\tilde{D}+L)^{-1} A (\tilde{D}+L)^{-t}] [(\tilde{D}+L)^t x] = [(\tilde{D}+L)^{-1} b]$$

or

$$(5) \quad \hat{A} \hat{x} = \hat{b}.$$

But applying PCG to (1) with  $M = (\tilde{D}+L)\tilde{D}^{-1}(\tilde{D}+L)^t$  is equivalent to applying PCG to (5) with  $\hat{M} = \tilde{D}^{-1}$  and setting  $x = (\tilde{D}+L)^{-t}\hat{x}$ .<sup>4</sup> If we update  $x$  instead of  $\hat{x}$  at each iteration, algorithm (2) becomes:

$$(6a) \quad \hat{p}_0 = \hat{r}_0 = \hat{b} - \hat{A}x_0$$

$$(6b) \quad \text{Compute } \hat{r}'_0 = \tilde{D}\hat{r}_0$$

FOR  $k = 0$  STEP 1 UNTIL Convergence DO

$$(6c) \quad \hat{a}_k = (\hat{r}_k, \hat{r}'_k) / (\hat{p}_k, \hat{A}\hat{p}_k)$$

$$(6d) \quad x_{k+1} = x_k + \hat{a}_k (\tilde{D}+L)^{-t} \hat{p}_k$$

$$(6e) \quad \hat{r}_{k+1} = \hat{r}_k - \hat{a}_k \hat{A}\hat{p}_k$$

$$(6g) \quad \text{Compute } \hat{r}'_{k+1} = \tilde{D}\hat{r}_{k+1}$$

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<sup>4</sup> Both are equivalent to applying the basic conjugate gradient method to the preconditioned system

$$\bar{A}\bar{x} = [\tilde{D}^{1/2}(\tilde{D}+L)^{-1} A (\tilde{D}+L)^{-t}\tilde{D}^{1/2}] [\tilde{D}^{-1/2}(\tilde{D}+L)^t x] = [\tilde{D}^{1/2}(\tilde{D}+L)^{-1} b] = \bar{b}$$

(see [4], pp. 58-59).

$$(6g) \quad \hat{b}_k = (\hat{r}_{k+1}, \hat{r}'_{k+1}) / (\hat{r}_k, \hat{r}'_k)$$

$$(6h) \quad \hat{p}_{k+1} = \hat{r}'_{k+1} + \hat{b}_k \hat{p}_k$$

$\hat{A}\hat{p}_k$  can be computed efficiently by taking advantage of the following identity:

$$(7) \quad \begin{aligned} \hat{A}\hat{p}_k &= (\tilde{D}+L)^{-1} [(\tilde{D}+L) + (\tilde{D}+L)^t - (2\tilde{D}-D)] (\tilde{D}+L)^{-t} \hat{p}_k \\ &= (\tilde{D}+L)^{-t} \hat{p}_k + (\tilde{D}+L)^{-1} [\hat{p}_k - K(\tilde{D}+L)^{-t} \hat{p}_k] , \end{aligned}$$

where  $K = 2\tilde{D}-D$ . Thus

$$(8a) \quad \hat{t}_k = (\tilde{D}+L)^{-t} \hat{p}_k$$

$$(8b) \quad \hat{A}\hat{p}_k = \hat{t}_k + (\tilde{D}+L)^{-1} (\hat{p}_k - K\hat{t}_k) ,$$

which requires  $2N+NZ(A)$  multiply-adds.  $\hat{t}_k$  can also be used to update  $x_k$  in (6d), so that the total cost for each PCG iteration is just  $8N+NZ(A)$  multiply-adds,<sup>5</sup> versus  $6N+2NZ(A)$  for the straight-forward implementation.

### 3. Generalizations

The approach presented in Section 2 extends immediately to preconditionings of the form

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<sup>5</sup> Again,  $3N$  multiply-adds can be saved by symmetrically scaling the problem so that  $\tilde{D} = I$ .



$$(9) \quad M = (\tilde{D}+L) \tilde{S}^{-1} (\tilde{D}+L)^t,$$

where  $\tilde{S}$  is positive diagonal. Moreover, if we take  $K = \tilde{D} + \tilde{D}^t - D$  in (7) and (8), then  $\tilde{D}$  need not be diagonal or even symmetric. In this case,  $\tilde{D}$  would reflect changes to both the diagonal and off-diagonal entries of  $A$  in generating an incomplete factorization. If we assume that only the nonzero entries of  $A$  are changed, i.e., that  $(K)_{ij}$  is nonzero only if  $(A)_{ij}$  is nonzero, then the operation count is  $7N + NZ(A) + NZ(K)$ .

Another application is to preconditioning nonsymmetric systems. Let

$$(10) \quad M = (\tilde{D}+L) \tilde{S}^{-1} (\tilde{D}+U),$$

be an incomplete LDU factorization of a nonsymmetric matrix  $A$ , where  $A = L + D + U$ ,  $L$  (respectively,  $U$ ) is strictly lower (respectively, upper) triangular, and  $D$  and  $\tilde{S}$  are diagonal. Then a number of authors have proposed solving the linear system  $Ax = b$  by solving the normal equations for one of the preconditioned systems

$$(11a) \quad \hat{A}_1 \hat{x} = [\tilde{S} (\tilde{D}+L)^{-1} A (\tilde{D}+U)^{-1}] [(\tilde{D}+U) x] = [\tilde{S} (\tilde{D}+L)^{-1} b] = \hat{b}$$

(see [12]) and

$$(11b) \quad \hat{A}_2 x = [(\tilde{D}+U)^{-1} \tilde{S} (\tilde{D}+L)^{-1} A] x = [(\tilde{D}+U)^{-1} \tilde{S} (\tilde{D}+L)^{-1} b] = \hat{b}$$

(see [14, 3]).  $\hat{A}_2 \hat{p}$  can be computed as

$$(12) \quad \hat{A}_2 \hat{p} = (\tilde{D}+U)^{-1} \tilde{S} [\hat{p} + (\tilde{D}+L)^{-1} (D+U-\tilde{D}) \hat{p}]$$

in  $4N + NZ(L) + 2NZ(U)$  multiply-adds, whereas  $\hat{A}_1 \hat{p}$  can be computed as

$$(13a) \quad \hat{t} = (\tilde{D}+U)^{-1} \hat{p}$$

$$(13b) \quad \hat{A}_1 \hat{p} = \tilde{S} [\hat{t} + (\tilde{D}+L)^{-1} (\hat{p} - (2\tilde{D}-D)\hat{t})]$$

in  $4N+NZ(L)+NZ(U)$  multiply-adds. Thus the first approach would be more efficient per iteration, although more iterations might be required to achieve comparable accuracy [14].<sup>6</sup>

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<sup>6</sup> The same would be true if a Generalized Conjugate Residual method such as Orthomin [15, 8] were used to solve (11a) or (11b).

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